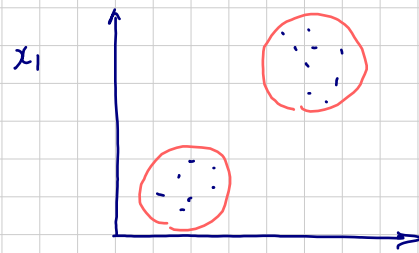


# Machine Learning Week 8

## Unsupervised Learning

### 1. Clustering



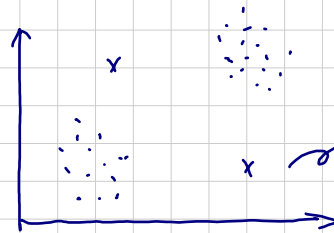
group / clustering:  
find some structures

Training set  $\{x^{(1)}, x^{(2)}, \dots, x^{(n)}\}$

#### Applications

- market segmentation
- social network analysis
- organize computer clusters
- astronomical data analysis

### 2. K-means Algorithm



cluster centroids. (randomly initiated)

#### step (1) cluster assignment.

- binary assignment of datasets. depending on proximity.

#### (2) Move centroid.

- move to "mean" of location in all labelled

re-colour.

[Input] .

①  $k$  (number of clusters)

② Training set  $\{x^{(1)}, x^{(2)}, \dots, x^{(n)}\}$

$x^{(i)} \in \mathbb{R}^n$  [drop  $x_0 = 1$  convention]

[Algorithm]

Randomly initialize  $K$  cluster centroids  $\mu_k, k=1 \dots K$

do  $\{$   
  for  $i=1 \dots n$ .

cluster assignment (

$$c_i := \min_k \|x^{(i)} - \mu_k\|^2$$

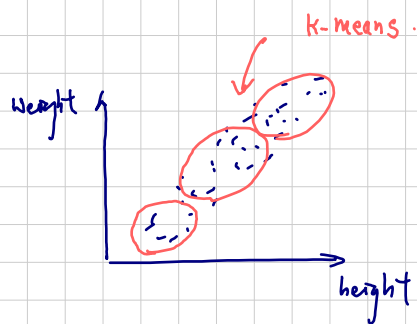
for  $k=1 \dots K$

move centroid (

$$\mu_k := \text{mean}(\text{pts assigned to cluster } k) \\ \in \mathbb{R}^n$$

3. K-means for non-separated clusters.

T-shirt song



## Optimization objective

K-means optimization objective.

- $c^i$  : index.
- $\mu_k$  : cluster centroid  $k \in \mathbb{R}^n$
- $\mu_{c^i}$  = cluster centroid of cluster to which  $x^{(i)}$  has been assigned.

$$J(c^{(1)}, \dots, c^{(m)}, \mu_1, \dots, \mu_K) = \frac{1}{m} \sum_{i=1}^m \|x^{(i)} - \mu_{c^{(i)}}\|^2$$



[Algorithm]

Randomly initialize  $K$  cluster centroids  $\mu_k$ ,  $k=1 \dots K$

do  $\downarrow$   
for  $i=1 \dots m$ .

cluster assignment  $\left( \begin{array}{l} c^i := \min_k \|x^{(i)} - \mu_k\|^2 \\ \text{for } k=1 \dots K \end{array} \right.$

move centroid  $\left( \begin{array}{l} \mu_k := \text{mean}(\text{pts assigned to cluster } k) \\ \in \mathbb{R}^n \end{array} \right.$

# Minimize  $J(\dots)$   
w.r.t.  $c^{(1)} \dots c^{(m)}$ .  
[holding  $\mu_1 \dots \mu_K$ ]

# Min. w.r.t.  $\mu_1 \dots \mu_K$

## Random Initialization

Rules.

- (1)  $K < m$
- (2) Randomly pick  $K$  training examples
- (3) set  $\mu_k = \text{examples}$

• Might have different clustering (local optimum)  $\rightarrow$  try different random initialization.

- Implementation.

for  $i = 1 \dots 1000$  {

    randomly initialize  $K$ -means

    run  $K$ -means. get  $c^{(1)} \dots c^{(K)}, \mu_1 \dots \mu_K$

    complete cost function (distortion)

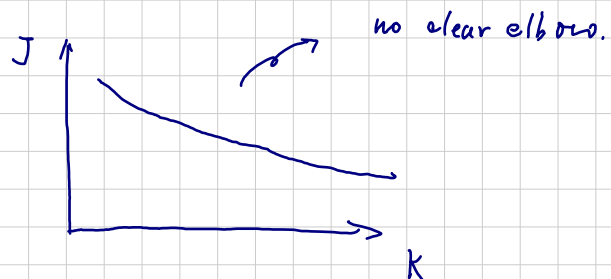
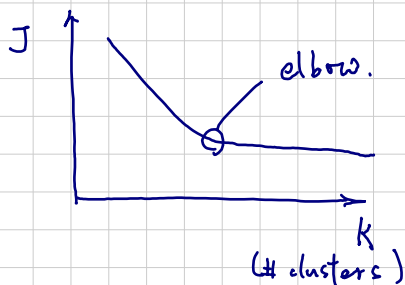
$\hookrightarrow J(c^{(1)}, \dots, c^{(K)}, \mu_1 \dots \mu_K)$

}

Pick one with lowest cost  $J$ .

## Choose $K$

1. Elbow method:



2. Pre-defined (f-chart size: 2...5).

